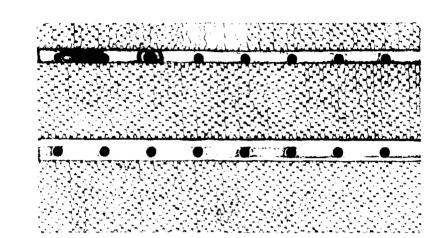


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EFFECT OF ADDITIONAL VARIABLES IN PRINCIPAL COMPONENT ANALYSIS, DISCRIMINANT ANALYSIS AND CANONICAL CORRELATION ANALYSIS*

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August 1985

Technical Report 85-31



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* Research sponsored by the Air Force Office of Scientific Research (AFSC), under contracts F49620-82-K-0001 and F49620-85-C-0008. This work was performed at the Center for Multivariate Analysis. The United States Government is authorized to reproduce and distribute reprints for governmental purposes notwithstanding any copyright notation hereon.

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1. INTRODUCTION

In a number of situations, it is of interest to find out whether the addition of a new set of variables gives additional information for inference. For example, in the area of principal component analysis, it is of interest to find out whether the new variables contribute to explanation of the variation among experimental units. In the area of multi-group discriminant analysis, it is important to find out whether the addition of new variables contributes to the discrimination between the groups. Similarly, in the area of canonical correlation analysis, it is of interest to find out as to whether the addition of variables to one or both sets of variables contributes to the degree of association between the two sets of variables.

Rao (1966) considered the effect of additional variables on the efficiency of estimates and the power of the test under multivariate regression model. Recently, Wijsman (1984) derived asymptotic distribution of the increase in the largest sample canonical correlation whem some variables are added. In Section 3 of this paper, we first derive asymptotic distributions of changes in functions of the eigenvalues of the sample covariance matrix. Asymptotic distributions of changes in functions of the eigenvalues of the multivariate analysis of variance (MANOVA) matrix when some variables are added are derived in Section 4. In Section 5, we derive asymptotic distributions of changes in certain functions of the sample canonical correlations when new variables are added to one or both sets of original variables. The above results are derived under the assumption that the underlying distribution is multivariate normal. Further results are given in Section 6.

2. PRELIMINARIES

In this section, we state the three lemmas which are needed in the sequel. Lemma 2.1. [Cramér (1946, p.366)]. Let x_n be a p-component random vector and μ : $p \times 1$ be a fixed vector. Assume \sqrt{n} $(x_n - u)$ converges to $N(0, \Sigma)$ in law. Let f(x) be continuously differentiable in a neighborhood of μ and let $\xi = \partial f(x)/\partial x \big|_{x=\mu}$. Then the limiting distribution is the same as the one of $\sqrt{n} \xi'(x_n - \mu)$, which is $N(0, \xi' \Sigma \xi)$.

Lemma 2.2. Let nS be distributed as a Wishart distribution $W_p(\Sigma,n)$ and B be distributed as a noncentral Wishart distribution $W_p(\Sigma,q;\Omega)$. Assume $\Omega=0(n)=n\theta$. Then the limiting distributions of $V=\sqrt{n}$ (S - Σ) and $U=\sqrt{n}$ ($\frac{1}{n}$ B - θ) are multivariate normal. Further the limiting distributions of trCV and trCU are $N(0,\sigma_1^2)$ and $N(0,\sigma_2^2)$ respectively, where C is a symmetric matrix of order $p\times p$, $\sigma_1^2=2$ tr(C Σ) and $\sigma_2^2=4$ tr C² θ .

This lemma is well known and is proved by considering the ch. functions of $\, V \,$ and $\, U \,$.

Lemma 2.3. Let $S_{1,n}$ and $S_{2,n}$ be sequences of symmetric matrices of order $p \times p$ such that the limiting distributions of $V_{1,n} = \sqrt{n}$ $(S_{1,n} - \Lambda)$ and $V_{2,n} = \sqrt{n}$ $(S_{2,n} - I_p)$ are multivariate normal, where $\Lambda = \operatorname{diag}(\lambda_1, \ldots, \lambda_p)$, $\lambda_1 \geq \ldots \geq \lambda_p$ and I_p is the identity matrix of order $p \times p$. Let $\ell_1 \geq \ldots \geq \ell_p$ and $\ell_2 \geq \ldots \geq \ell_p$ be the eigenvalues of $S_{1,n}$ and $S_{1,n}S_{2,n}^{-1}$, respectively. Suppose that the α -th largest eigenvalue $\ell_1 = \ell_1$ are the same as the ones of $\ell_1 = \ell_2$ and $\ell_2 = \ell_3$ and $\ell_3 = \ell_4$ are the same as the ones of $\ell_1 = \ell_4$ and $\ell_2 = \ell_4$ and $\ell_3 = \ell_4$ are the same denotes the $\ell_1 = \ell_4$ and $\ell_2 = \ell_4$ and $\ell_3 = \ell_4$ are the same denotes the $\ell_1 = \ell_4$ and $\ell_2 = \ell_4$ and $\ell_3 = \ell_4$ are the same denotes the $\ell_4 = \ell_4$ and $\ell_4 = \ell_4$ and $\ell_5 = \ell_4$ are the same denotes the $\ell_4 = \ell_4$ and $\ell_5 = \ell_4$ and $\ell_5 = \ell_4$ and $\ell_5 = \ell_4$ and $\ell_5 = \ell_4$ are the same as the ones of $\ell_5 = \ell_5$ and $\ell_5 = \ell_5$ and $\ell_5 = \ell_5$ are the same as the ones of $\ell_5 = \ell_5$ and $\ell_5 = \ell_5$ are the same as the ones of $\ell_5 = \ell_5$ and $\ell_5 = \ell_5$ and $\ell_5 = \ell_5$ and $\ell_5 = \ell_5$ are the same as the ones of $\ell_5 = \ell_5$ and $\ell_5 = \ell_5$ and $\ell_5 = \ell_5$ are the same as the ones of $\ell_5 = \ell_5$ and $\ell_5 = \ell_5$ are the same as the ones of $\ell_5 = \ell_5$ and $\ell_5 = \ell_5$ and $\ell_5 = \ell_5$ are the same as the ones of $\ell_5 = \ell_5$ and $\ell_5 = \ell_5$ are the same as the ones of $\ell_5 = \ell_5$ and $\ell_5 = \ell_5$ are the same as the ones of $\ell_5 = \ell_5$ and $\ell_5 = \ell_5$ and $\ell_5 = \ell_5$ are the same as the ones of $\ell_5 = \ell_5$ and $\ell_5 = \ell_5$ are the same as the ones of $\ell_5 = \ell_5$ and $\ell_5 = \ell_5$ are the same as the ones of $\ell_5 = \ell_5$ and $\ell_5 = \ell_5$ are the same as the ones of $\ell_5 = \ell_5$ and $\ell_5 = \ell_5$ are the same as the ones of $\ell_5 = \ell_5$ and $\ell_5 = \ell_5$ are the same as the ones of $\ell_5 = \ell_5$ and $\ell_5 = \ell_5$ are the same as the ones of $\ell_5 = \ell_5$ and $\ell_5 = \ell_5$ are the same as the ones of $\ell_5 = \ell_5$ and $\ell_5 = \ell_5$ are the same as the ones of $\ell_5 = \ell_5$ and ℓ_5

This lemma has been essentially proved in the papers of Hsu (1941a,b) and Anderson (1963) who treated the general case of multiple roots.

3. PRINCIPAL COMPONENT ANALYSIS

In the area of principal component analysis, it is of interest to find out a small number of principal components which would adequately explain the variation among experimental units. In the population, the variance of i-th important principal component is the i-th largest eigenvalue of the population covariance matrix. If these eigenvalues are small, then the corresponding principal components are unimportant. In a number of situations, it is of interest to find out as to whether the addition of some variables will increase the variances of the first few important principal components. Similarly, it is of interest to find out whether there is significant increase in the ratio of the i-th largest eigenvalue to the trace of the covariance matrix if some variables are added. So, we will derive asymptotic distributions of increases in certain functions of the eigenvalues of the sample covariance matrix when a new set of variables is added.

Let x: p × 1 be distributed as N(0, Σ). We partition x = (x_1', x_2'), x_1 : $p_1 \times 1$ and

$$\Sigma = \begin{bmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{bmatrix}, \Sigma_{11} : P_1 \times P_1.$$
 (3.1)

Suppose the vector \mathbf{x}_1 is augmented to \mathbf{x} . Let λ_{α} and $\tilde{\lambda}_{\alpha}$ be the α -th largest roots of Σ_{11} and Σ , respectively. Then

$$\tilde{\lambda}_{\alpha} \geq \lambda_{\alpha} \ (\alpha = 1, \ldots, p_1)$$

which follows from the Poincaré separation theorem (see, e.g., Rao (1973, p.64)).

We are interested in the increases $\delta_{\alpha} = \lambda_{\alpha} - \lambda_{\alpha}$. Let S be the sample covariance matrix based on a sample of size N = N + 1. We partition S as in (3.1)

$$S = \begin{bmatrix} S_{11} & S_{12} \\ S_{21} & S_{22} \end{bmatrix} . \tag{3.2}$$

The sample quantities corresponding to δ_{lpha} are

$$d_{\alpha} = \tilde{\ell}_{\alpha} - \ell_{\alpha}, (\alpha = 1, \ldots, p_1)$$

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where ℓ_α and $\tilde{\ell}_\alpha$ are the α -th largest roots of S_{11} and S, respectively. We consider the distribution of

$$J = \sqrt{n} \{ f(d_1, ..., d_{p_1}) - f(\delta_1, ..., \delta_{p_1}) \}$$
 (3.3)

We assume

Al: f(d) is continuously differentiable in a neighborhood of $d = \delta$

where $d = (d_1, ..., d_{p_1})'$ and $\delta = (\delta_1, ..., \delta_{p_1})'$. Let

$$c = (c_1, \dots, c_{p_1})' = \frac{\partial}{\partial d} f(d) \Big|_{d = \delta}$$
(3.4)

Let $H_{11}: p_1 \times p_1$ be an orthogonal matrix such that

$$H_{11}^{\Sigma}_{11}H_{11}' = \Lambda_{11} = diag(\lambda_1, ..., \lambda_{p_1}).$$

Since l_{α} and \tilde{l}_{α} are invariant under the transformation

$$x \longrightarrow \begin{bmatrix} H_{11} & 0 \\ 0 & I \end{bmatrix} x.$$

we may assume

$$\Sigma = \Lambda$$

$$= \begin{bmatrix} \Lambda_{11} & \Lambda_{12} \\ & & \\ \Lambda_{21} & \Lambda_{22} \end{bmatrix}$$
(3.5)

where
$$\Lambda_{12} = H_{11}\Sigma_{12}$$
, $\Lambda_{21} = \Sigma_{21}H_{11}^{*}$ and $\Lambda_{22} = \Sigma_{22}$. Let
$$S = \Lambda + \frac{1}{\sqrt{n}}V$$
(3.6)

and

$$\tilde{S} = \Gamma' S \Gamma$$

$$= \tilde{\Lambda} + \frac{1}{\sqrt{n}} \Gamma' V \Gamma$$
(3.7)





where Γ is an orthogonal matrix such that $\Gamma'\Lambda\Gamma = \tilde{\Lambda} = \mathrm{diag}\ (\tilde{\lambda}_1,\dots,\tilde{\lambda}_p)$. Since ℓ_α and $\tilde{\ell}_\alpha$ are the α -th largest roots of $S_{11} = \Lambda_{11} + (1/\sqrt{n}) \ V_{11}$ and $\tilde{S} = \tilde{\Lambda} + (1/\sqrt{n}) \ \Gamma'V\Gamma$, we obtain by Lemma 2.3 that the asymptotic distribution of \sqrt{n} $(d_\alpha - \delta_\alpha)$ is the same as that of

$$g_{\alpha} = (\Gamma'V\Gamma)_{\alpha\alpha} - (V)_{\alpha\alpha}$$
 (3.8)

if λ_{α} and $\tilde{\lambda}_{\alpha}$ are simple. Using Lemma 2.1 we obtain that the asymptotic distribution of J is the same as that of

$$\sum_{\alpha=1}^{p_1} c_{\alpha} g_{\alpha} = \text{tr A V}$$
 (3.9)

where

$$A = \Gamma D_c \Gamma' - D_c, D_c = diag(c_1, ..., c_{p_1}, 0, ..., 0).$$
 (3.10)

This implies the following.

Theorem 3.1. Let nS be distributed as a Wishart distribution $W_p(\Sigma,n)$. Let $d_\alpha = \tilde{\ell}_\alpha - \ell_\alpha$ and $\delta_\alpha = \tilde{\lambda}_\alpha - \lambda_\alpha$, where ℓ_α , $\tilde{\ell}_\alpha$, λ_α and $\tilde{\lambda}_\alpha$ are the α -th largest roots of S_{11} , S, Σ_{11} and Σ . Assume a function $f(d_1,\ldots,d_{p_1})$ satisfies the assumption Al and all the roots λ_α and $\tilde{\lambda}_\alpha$ ($\alpha=1,\ldots,p_1$) are simple. Then

$$\sqrt{n} \{f(d_1,...,d_{p_1}) - f(\delta_1,...,\delta_{p_1})\} \xrightarrow{D} N(0,\sigma^2)$$

as $n \rightarrow \infty$, where

$$\sigma^{2} = 2 \operatorname{tr} (A\Lambda)^{2}$$

$$= 2 \sum_{\alpha=1}^{p_{1}} c_{\alpha}^{2} (\lambda_{\alpha}^{2} + \tilde{\lambda}_{\alpha}^{2}) - 4 \sum_{\alpha, \beta=1}^{p_{1}} c_{\alpha} c_{\beta} \tilde{\lambda}_{\alpha}^{2} \gamma_{\beta\alpha}^{2}$$

and $\Gamma = (\gamma_{\alpha\beta})$.

Cor. 3.1.1. When λ_{α} and $\tilde{\lambda}_{\alpha}$ are simple,

$$\sqrt{n} \left\{ (\tilde{\ell}_{\alpha} - \ell_{\alpha}) - (\tilde{\lambda}_{\alpha} - \lambda_{\alpha}) \right\} \xrightarrow{D} N(0, \sigma^{2})$$

as $n \to \infty$, where $\sigma^2 = 2(\lambda^2 + \tilde{\lambda}^2) - 4 \tilde{\lambda}_{\alpha}^2 \gamma_{\alpha\alpha}^2$.

4. EFFECT OF ADDITIONAL VARIABLES IN DISCRIMINANT ANALYSIS

In the area of discriminant analysis, it is of interest to find out as to whether the addition of a new set of variables will make a significant contribution on the discriminant functions. This problem can be investigated by examining the increases due to the additional variables in certain functions of the eigenvalues of the MANOVA matrix. So, we will study the asymptotic distributions of the above increases in the sample.

Let W and B be independently distributed as a Wishart distribution $W_p(\Sigma,n) \text{ and a noncentral Wishart distribution } W_p(\Sigma,q:\Xi)$ We partition

$$W = \begin{pmatrix} W_{11} & W_{12} \\ W_{21} & W_{22} \end{pmatrix}, \qquad B = \begin{pmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{pmatrix},$$

$$\Sigma = \begin{pmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{pmatrix}, \qquad \Xi = \begin{pmatrix} \Xi_{11} & \Xi_{12} \\ \Xi_{21} & \Xi_{22} \end{pmatrix}, \qquad (4.1)$$

with W_{11} : $p_1 \times p_1$, B_{11} : $p_1 \times p_1$, Σ_{11} : $p_1 \times p_1$ and Ξ_{11} : $p_1 \times p_1$. Let ℓ_{α} , ℓ_{α} , ω_{α} and ω_{α} be the α -th largest roots of $B_{11}W_{11}^{-1}$, BW_{11}^{-1} , BW_{11}^{-1} , $E_{11}\Sigma_{11}^{-1}$ and E_{Σ}^{-1} , respectively. Then

$$d_{\alpha} = \hat{l}_{\alpha} - l_{\alpha} \ge 0,$$

$$n\delta_{\alpha} = \hat{\omega}_{\alpha} - \omega_{\alpha} \ge 0, \quad (\alpha = 1, \dots, p_{1})$$
(4.2)

which follows from the Poincaré separation theorem in the case of two matrices (also see Gabriel (1968)).

Let

where L Σ L' = I and H₁₁ is an orthogonal matrix such that H₁₁L₁₁E₁₁L'₁₁H'₁₁ = Ω_{11} = diag($\omega_1,\ldots,\omega_{p_1}$). Since ℓ_α and $\tilde{\ell}_\alpha$ are invariant under the transformation B \longrightarrow TBT' and W \longrightarrow TWT', we may assume

$$\Xi = \Omega = \begin{bmatrix} \Omega_{11} & \Omega_{12} \\ & & \\ \Omega_{21} & \Omega_{22} \end{bmatrix}$$
 (4.4)

with $\Omega_{11} = \text{diag}(\omega_1, \dots, \omega_{p_1})$, $\Omega_{12} = \text{H}_{11} \text{L}_{11} (\Xi_{11} \text{L}_{21}^{\dagger} + \Xi_{12} \text{L}_{22}^{\dagger})$, $\Omega_{21} = \Omega_{12}^{\dagger}$ and $\Omega_{22} = (\text{L}_{21}\Xi_{11} + \text{L}_{22}\Xi_{21}) \text{L}_{21}^{\dagger} + (\text{L}_{21}\Xi_{12} + \text{L}_{22}\Xi_{22}) \text{L}_{22}^{\dagger}$. We assume

A2:
$$\Omega = O(n)$$

$$= n \Theta = n \begin{bmatrix} \theta & \theta & \theta \\ 11 & \theta & 12 \\ \theta & \theta & \theta \\ 21 & \theta & 22 \end{bmatrix}$$
 (4.5)

where $\theta_{11} = diag(\theta_1, \dots, \theta_{p_1})$. Let Γ be an orthogonal matrix such that

$$\Gamma' \Theta \Gamma = \widetilde{\Theta} = \operatorname{diag}(\widetilde{\theta}_1, \dots, \widetilde{\theta}_p)$$
 (4.6)

with $\tilde{\theta}_1 \geq \ldots \geq \tilde{\theta}_p$. Then $n\tilde{\theta}_{\alpha} = \tilde{\omega}_{\alpha}$. Let

$$\frac{1}{n}B = 6 + \frac{1}{\sqrt{n}} U,$$

$$\frac{1}{n}W = I + \frac{1}{\sqrt{n}} W.$$
(4.7)

Then it is easily seen that ℓ_{α} and ℓ_{α} are the α -th largest roots of

$$|\Theta_{11} + \frac{1}{\sqrt{2}} U_{11} - \ell(I - \frac{1}{\sqrt{2}} V_{11})| = 0$$

and

$$|\tilde{\theta}| + \frac{1}{\sqrt{n}} \Gamma'U\Gamma - \tilde{\ell}(I + \frac{1}{\sqrt{n}} \Gamma'V\Gamma)| = 0$$

respectively, where \mathbf{U}_{11} : $\mathbf{p}_1 \times \mathbf{p}_1$ and \mathbf{V}_{11} : $\mathbf{p}_1 \times \mathbf{p}_2$ are the submatrices of U and V partitioned as in (4.1).

From Lemma 2.3, it is seen that the asymptotic distribution of \sqrt{n} $(d_{\alpha} - \delta_{\alpha})$ is the same as that of

$$g_{\alpha} = (\Gamma^{\dagger} U \Gamma)_{\alpha \alpha} - \tilde{\theta}_{\alpha} (\Gamma^{\dagger} V \Gamma)_{\alpha \alpha} - (V)_{\alpha \alpha} + \theta_{\alpha} (V)_{\alpha \alpha}$$
 (4.8)

if ω_{α} and $\widetilde{\omega}_{\alpha}$ are simple. Using Lemma 1, we obtain that the asymptomatic distribution of \sqrt{n} $\{f(d_1,\ldots,d_{p_1}) - f(\delta_1,\ldots,\delta_{p_1})\}$ is the same as that of

$$\sum_{\alpha=1}^{p_1} c_{\alpha} g_{\alpha} = trA^{(1)}V + trA^{(2)}U$$
 (4.9)

where $A^{(1)} = D_{c\theta} - D_{c\theta} \Gamma^{\dagger}$, $A^{(2)} = \Gamma D_{c} \Gamma^{\dagger} - D_{c}$, $D_{c} = diag(c_{1},...,c_{p_{1}},0,...,0)$,

 $D_{c\theta} = diag(c_1\theta_1,...,c_{p_1},0,...,0)$, etc. This implies the following.

Theorem 4.1. Let W and B be independently distributed as a Wishart distribution $W_p(\Sigma,q;\Xi)$, respectively. Let $d_\alpha=\hat{\ell}_\alpha-\ell_\alpha$ and $n\delta_\alpha=\hat{\omega}_\alpha-\omega_\alpha$, where ℓ_α , ℓ_α , ω_α and ω_α are the α -th largest roots of $B_{11}W_{11}^{-1}$, BW^{-1} , $\Xi_{11}\Sigma_{11}^{-1}$ and $\Xi\Sigma^{-1}$, respectively. Assume that the assumptions Al and A2 are satisfied, and ω_α and ω_α ($\alpha=1,\ldots,p_1$) are simple. Then

$$\sqrt{n} \{f(d_1, ..., d_{p_1}) - f(\delta_1, ..., \delta_{p_1})\} \xrightarrow{D} N(0, \sigma^2)$$

as $n \rightarrow \infty$, where

$$\sigma^{2} = 2 \operatorname{tr} \{ D_{c\theta} - \Gamma D_{c\tilde{\theta}} \Gamma' \}^{2} + 4 \operatorname{tr} \{ \Gamma D_{c} \Gamma' - D_{c} \}^{2} \Theta$$

$$= 2 \sum_{\alpha=1}^{p_{1}} c_{\alpha}^{2} \{ \theta_{\alpha} (\theta_{\alpha} + 2) + \tilde{\theta}_{\alpha} (\tilde{\theta}_{\alpha} + 2)$$

$$- 4 \sum_{\alpha, \beta=1}^{p_{1}} c_{\alpha} c_{\beta} \tilde{\theta}_{\beta} \gamma_{\alpha\beta}^{2} (\theta_{\alpha} + 2)$$

and $\Gamma = (\gamma_{\alpha\beta})$.

Cor. 4.1.1. When ω_{α} and $\tilde{\omega}_{\alpha}$ are simple

$$\sqrt{n} \ [(\tilde{\ell}_{\alpha} - \ell_{\alpha}) - (\tilde{\theta}_{\alpha} - \theta_{\alpha})] \xrightarrow{D} N(0,\sigma^{2})$$

as
$$n \to \infty$$
, where $\sigma^2 = 2 \theta_{\alpha}(\theta_{\alpha} + 2) + 2 \tilde{\theta}_{\alpha}(\tilde{\theta}_{\alpha} + 2) - 4 \tilde{\theta}_{\alpha}(\theta_{\alpha} + 2)\gamma_{\alpha\alpha}^2$.

5. EFFECT OF ADDITIONAL VARIABLES ON CANONICAL VARIABLES

In this section, we study asymptotic distributions of certain statistics useful in studying the effect of additional variables on the canonical correlations.

Consider two sets of variables $x_1 \colon p_1 \times 1$ and $y_1 \colon q_1 \times 1$. We assume $p_1 \leq q_1$. Let $p_1 \geq \dots \geq p_{p_1} \geq 0$ be the canonical correlations between x_1 and y_1 . We shall augment the variates x_1 and y_1 to $x \colon p \times 1$ and $y \colon q \times 1$ by adding extra variates $x_2 \colon p_2 \times 1$ and $y_2 \colon q_2 \times 1$, respectively. We assume that (x',y') is distributed as $N_{p+q}[\mu,\Sigma]$. We partition Σ as

$$\Sigma = \begin{pmatrix} \Sigma_{xx} & \Sigma_{xy} \\ & & \\ \Sigma_{yx} & \Sigma_{yy} \end{pmatrix} , \Sigma_{xx} : p \times p.$$
 (5.1)

Let ρ_{α} be the $\alpha-th$ largest canonical correlation between x and y. Then

$$\delta_{\alpha} = \rho_{\alpha} - \rho_{\alpha} \ge 0 \quad (\alpha = 1, ..., p_1). \tag{5.2}$$
 which has been shown in Fujikoshi (1982).
Let

$$S = \begin{bmatrix} S_{xx} & S_{xy} \\ S_{yx} & S_{yy} \end{bmatrix}$$

be the sample covariance matrix besed on a sample of size N = n + 1. Let r_{α} and r_{α} be the α -th largest canonical correlations between x_{α} and y_{α} and between x_{α} and y_{α} . We consider the asymptotic distribution of

$$\sqrt{n} \{f(d_1,\ldots,d_{p_1}) - f(\delta_1,\ldots,\delta_{p_1})\}$$

where $d_{\alpha} = r_{\alpha} - r_{\alpha}$. Let L_1 and L_2 be the lower triangular matrices such that

$$L_1^{\Sigma}_{xx}L_1^{\dagger} = I_p, L_2^{\Sigma}_{yy}L_2^{\dagger} = I_q.$$
 (5.3)

We partition

$$L_{1} = \begin{pmatrix} L_{1 \cdot 11} & 0 \\ L_{1 \cdot 21} & L_{1 \cdot 22} \end{pmatrix}, \quad L_{2} = \begin{pmatrix} L_{2 \cdot 11} & 0 \\ L_{2 \cdot 21} & L_{2 \cdot 22} \end{pmatrix}.$$
(5.4)

Let $H_{1\cdot 11}$: $p_1 \times p_1$ and $H_{2\cdot 11}$: $p_2 \times p_2$ be the orthogonal matrices chosen so that

$$H_{1\cdot11}^{L_{1\cdot11}} xy \cdot 11^{L_{2\cdot11}^{I}} = P_{11}$$

$$= \begin{pmatrix} \rho_{1} & & & \\ & \ddots & & \\ & & \rho_{1} & & \end{pmatrix}, \qquad (5.5)$$

where

$$\Sigma_{xy} = \begin{bmatrix} \Sigma_{xy*11} & \Sigma_{xy*12} \\ \Sigma_{xy*21} & \Sigma_{xy*22} \end{bmatrix}, \Sigma_{xy*11} : p_1 \times q_1.$$
 (5.6)

Then r_{α} and r_{α} are invariant under the transformation

Therefore, we may assume

$$\Sigma_{xx} = I_{p}, \Sigma_{yy} = I_{q}$$

$$\Sigma_{xy} = P$$

$$= \begin{pmatrix} P_{11} & P_{12} \\ P_{21} & P_{22} \end{pmatrix}$$
(5.7)

with
$$P_{11}$$
 in (5.5), $P_{12} = H_{1 \cdot 11}L_{1 \cdot 11}$ ($\Sigma_{11}L_{2 \cdot 21}^{\dagger} + \Sigma_{12}L_{2 \cdot 22}^{\dagger}$),

$$P_{21} = (L_{1 \cdot 21}\Sigma_{11} + L_{1 \cdot 22}\Sigma_{21}) L_{2 \cdot 11}^{\dagger} H_{2 \cdot 11}^{\dagger} \quad \text{and} \quad P_{22} = (L_{1 \cdot 21}\Sigma_{11} + L_{1 \cdot 22}\Sigma_{21}) L_{2 \cdot 21}^{\dagger}$$

$$+ (L_{1 \cdot 21}\Sigma_{12} + L_{1 \cdot 22}\Sigma_{22}) L_{2 \cdot 22}^{\dagger} \quad \text{Let}$$

$$S = \Sigma + \frac{1}{\sqrt{n}} \quad V$$

$$= \begin{pmatrix} I_{p} & P \\ P' & I_{0} \end{pmatrix} + \frac{1}{\sqrt{n}} \begin{pmatrix} V_{xx} & V_{xy} \\ V_{yx} & V_{yy} \end{pmatrix} . \quad (5.8)$$

Then r_{α} is the α -th root of

$$|s_{xy*11}s_{yy*11}^{-1}s_{yx*11} - r s_{xx*11}| = 0$$

which is equivalent to

$$P_{11}P_{11}' + \frac{1}{\sqrt{n}}Z_{11} - r^{2}(I_{p_{1}} + \frac{1}{\sqrt{n}}V_{xx\cdot 11})| = 0$$
 (5.9)

where

$$Z_{11} = \sqrt{n} \{ (P_{11} + \frac{1}{\sqrt{n}} \nabla_{xy+11}) (I_{q_1} + \frac{1}{\sqrt{n}} \nabla_{y+11})^{-1} (P_{11} + \frac{1}{\sqrt{n}} \nabla_{yx+11}) - P_{11} P_{11} \}, (5.10)$$

 $S_{xy\cdot 11}$, $V_{xy\cdot 11}$, etc. denote the submatrices of S_{xy} and V_{xy} partitioned as in (5.6). Let Γ_1 and Γ_2 be the orthogonal matrices such that

$$\Gamma_1 P \Gamma_2^* = \tilde{P} : p \times q \tag{5.11}$$

where the (α,α) elements of P are ρ_{α} and other elements are zero. Let

$$\tilde{\mathbf{S}} = \begin{bmatrix} \Gamma_1 & 0 \\ 0 & \Gamma_2 \end{bmatrix} \mathbf{S} \begin{bmatrix} \Gamma_1 & 0 \\ 0 & \Gamma_2 \end{bmatrix} = \begin{bmatrix} \Gamma_1 \mathbf{S}_{\mathbf{x}} \Gamma_1^{\prime} & \Gamma_1 \mathbf{S}_{\mathbf{x}} \Gamma_2^{\prime} \\ \Gamma_2 \mathbf{S}_{\mathbf{y}} \Gamma_2^{\prime} & \Gamma_2 \mathbf{S}_{\mathbf{y}} \Gamma_2^{\prime} \end{bmatrix}$$

$$= \begin{bmatrix} \mathbf{I}_{\mathbf{p}} & \mathbf{P} \\ \mathbf{P}^{\prime} & \mathbf{I}_{\mathbf{q}} \end{bmatrix} + \frac{1}{\sqrt{\mathbf{n}}} \begin{bmatrix} \Gamma_1 \mathbf{V}_{\mathbf{x}} \Gamma_1^{\prime} & \Gamma_1 \mathbf{V}_{\mathbf{x}} \Gamma_2^{\prime} \\ \Gamma_2 \mathbf{V}_{\mathbf{m}} \Gamma_1^{\prime} & \Gamma_2 \mathbf{V}_{\mathbf{m}} \Gamma_2^{\prime} \end{bmatrix} . \tag{5.12}$$

From (5.12) we obtain that r_{α} is the α -th largest root of

$$|\tilde{PP'} + \frac{1}{\sqrt{n}}\tilde{Z} - \tilde{r}^2(I + \frac{1}{\sqrt{n}}\Gamma_1 V_{xx}\Gamma_1')| = 0.$$
 (5.13)

where

$$\tilde{Z} = \sqrt{n} \{\tilde{P} + \frac{1}{\sqrt{n}} \Gamma_1 \nabla_{xx} \Gamma_1^{'}\} (I_q + \frac{1}{\sqrt{n}} \Gamma_2 \nabla_{yy} \Gamma_2^{'})^{-1} (\tilde{P} + \frac{1}{\sqrt{n}} \Gamma_2 \nabla_{yx} \Gamma_1^{'}) - \tilde{PP}^{'}\}$$
 (5.14)

We note that the limiting distributions of Z_{11} and \tilde{Z} are the same as those of $V_{xy}\cdot 11^P_{11} + P_{11}V_{yx}\cdot 11 - P_{11}V_{yy}\cdot 11^P_{11}$ and $\Gamma_1V_{xy}\Gamma_1^{'P'} + \tilde{P}\Gamma_2V_{yx}\Gamma_1^{'}$ $-\tilde{P}\Gamma_2V_{yy}\Gamma_2^{'P'}$, respectively. Applying Lemma 2.3 to (5.9) and (5.14) we obtain that the asymptotic distribution of \sqrt{n} $(d_{\alpha} - \delta_{\alpha})$ is the same as that of

$$g_{\alpha} = (\Gamma_{1} \nabla_{xy} \Gamma_{2}^{\dagger})_{\alpha\alpha} - (\nabla_{xy})_{\alpha\alpha}$$

$$-\frac{1}{2} \tilde{\rho}_{\alpha} (\Gamma_{1} \nabla_{xx} \Gamma_{1}^{\dagger})_{\alpha\alpha} - \frac{1}{2} \tilde{\rho}_{\alpha} (\Gamma_{2} \nabla_{yy} \Gamma_{2}^{\dagger})_{\alpha\alpha}$$

$$+\frac{1}{2} \tilde{\rho}_{\alpha} (\nabla_{xx})_{\alpha\alpha} + \frac{1}{2} \rho_{\alpha} (\nabla_{yy})_{\alpha\alpha}. \qquad (5.15)$$

Therefore the asymptotic distribution of \sqrt{n} $\{f(d_1,\ldots,d_{p_1}) - f(\delta_1,\ldots,\delta_{p_1})\}$ is the same as that of

$$\sum_{\alpha} c_{\alpha} g_{\alpha} = \frac{1}{2} \text{ tr AV}$$
 (5.16)

where

$$A = \begin{bmatrix} A_{11} & A_{12} \\ & & \\ A_{21} & A_{22} \end{bmatrix},$$

$$A_{11} = \operatorname{diag}(c_{1}^{\rho}_{1}, \dots, c_{p_{1}^{\rho}_{p_{1}}}, 0, \dots, 0) - \Gamma_{1}^{\prime} \operatorname{diag}(\tilde{c}_{1}^{\rho}_{1}, \dots, c_{p_{1}^{\rho}_{p_{1}}}, 0, \dots, 0)\Gamma_{1},$$

$$A_{22} = \operatorname{diag}(c_{1}^{\rho}_{1}, \dots, c_{p_{1}^{\rho}_{p_{1}}}, 0, \dots, 0) - \Gamma_{2}^{\prime} \operatorname{diag}(c_{1}^{\rho}_{1}, \dots, c_{p_{1}^{\rho}_{p_{1}}}, 0, \dots, 0)\Gamma_{2},$$

$$A_{21} = A_{12}^{\prime} = \Gamma_{2}^{\prime}D_{c} \ 1^{\Gamma-D_{c}}$$

and the (α,α) elements of D_c : $q \times p$ are c_{α} for $\alpha = 1, \ldots, p_1$ and the other elements of D_c are zero. This implies the following.

Theorem 5.1. Let r_{α} and r_{α} be the α -th largest canonical correlations between $x_1: p_1 \times 1$ and $y_1: q_1 \times 1 \ (p_1 \leq q_1)$ and between x: $p \times 1$ and y: $q \times 1$, based on a sample of size N = n + 1 from $N(\mu, \Sigma)$.

Let ρ_{α} and ρ_{α} be the corresponding population quantities. Assume that a function $f(d_1, \dots, d_{p_1})$ satisfies the assumption Al and the canonical correlations ρ_{α} and ρ_{α} (α = 1,..., ρ_{1}) are simple. Then

$$\sqrt{n} \left\{ f(d_1, \dots, d_{p_1}) - f(\delta_1, \dots, \delta_{p_1}) \right\} \xrightarrow{D} N(0, \sigma^2)$$
is $n \neq \infty$, where $d = r - r + \delta = 0$

as $n \rightarrow \infty$, where $d_{\alpha} = r_{\alpha} - r_{\alpha}$, $\delta_{\alpha} = \rho_{\alpha} - \rho_{\alpha}$,

$$\sigma^{2} = \frac{1}{2} \operatorname{tr} \left\{ \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \begin{bmatrix} I_{p} & P \\ P' & I_{q} \end{bmatrix} \right\}^{2}$$

$$= \sum_{\alpha, \beta=1}^{p_{1}} c_{\alpha}^{2} \left\{ (1 - \rho_{\alpha}^{2})^{2} + (1 - \tilde{\rho}_{\alpha}^{2})^{2} \right\}$$

$$+ \sum_{\alpha, \beta=1}^{p_{1}} c_{\alpha} c_{\beta} (1 - \tilde{\rho}_{\beta}^{2}) \left\{ \rho_{\alpha} \tilde{\rho}_{\beta} (\gamma_{1 \cdot \beta \alpha}^{2} + \gamma_{2 \cdot \beta \alpha}^{2}) - 2\gamma_{1 \cdot \beta \alpha} \gamma_{2 \cdot \beta \alpha} \right\}$$

and $\Gamma_1 = (\gamma_{1 \cdot \alpha \beta})$ and $\Gamma_2 = (\gamma_{2 \cdot \alpha \beta})$.

Cor. 5.1.1. When ρ_{α} and ρ_{α} are simple,

$$\sqrt{n} \left\{ (\tilde{r}_{\alpha} - r_{\alpha}) - (\tilde{\rho}_{\alpha} - \rho_{\alpha}) \right\} \xrightarrow{D} N(0, \sigma^{2})$$

as $n \to \infty$, where $\sigma^2 = (1 - \tilde{\rho}_{\alpha}^2)^2 + (1 - \tilde{\rho}_{\alpha}^2) \left[\tilde{\rho}_{\alpha} \tilde{\rho}_{\alpha} (\gamma_{1 \cdot \alpha \alpha}^2 + \gamma_{2 \cdot \alpha \alpha}^2) - 2 \gamma_{1 \cdot \alpha \alpha} \gamma_{2 \cdot \alpha \alpha}^2 \right]$

Note: Wijsman (1984) proved the Corr. 5.1.1 in the case of $\alpha = 1$ and $q_2 = 0 \text{ or } p_2 = 0.$

6. FURTHER RESULTS

Let d_{α} and δ_{α} be the increases in the α -th largest sample and population eigenvalues in the three cases: (i) principal component analysis, (ii) discriminant analysis, and (iii) canonical correlation analysis. Then in Theorems 3.1, 4.1, and 5.1, we have shown that

$$J = \sqrt{n} \{f(d_1, ..., d_{p_1}) - f(\sigma_1, ..., \sigma_{p_1})\} \stackrel{D}{+} N(0, \sigma^2).$$
 (6.1)

The limiting variance σ^2 depends on unknown Σ for cases (i) and (iii) and unknown Σ and Θ for case (ii). In order to make the formula useful, we need to estimate $\hat{\sigma}^2$ obtained from σ^2 by replacing Σ by S for (i) and (iii), and by replacing Σ and Θ by (1/n)W and (1/n)B, respectively, for (ii). It is easy to see that under the same assumption as in each of the Theorems

$$\hat{\sigma} \rightarrow \delta$$
 in probability.

Therefore, it follows from (6.1) that

$$\hat{\sigma}^{-1}J \stackrel{D}{+} N(0,1),$$
 (6.2)

the formula is useful in constructing an approximate confidence interval for $f(\delta_1,\dots,\delta_{p_1}).$

It is easy to extend the result for a single function J to the one for several functions. Let

$$J_{\alpha} = \sqrt{n} \left\{ f_{\alpha}(d_1, \dots, d_{p_1}) - f_{\alpha}(\delta_1, \dots, \delta_{p_1}) \right\}$$

for $\alpha = 1, 2, ..., k$. We assume that f_{α} 's satisfy the assumption Al. Let

$$c_{\alpha} = (c_{1,\alpha}, \dots, c_{p_1,\alpha})' = \frac{\partial}{\partial d} f_{\alpha}(d)$$

$$d = \delta$$

Then we can prove that

$$J = (J_1, ..., J_k) \stackrel{D}{\to} N_k(0, Q).$$
 (6.3)

The limiting covariance matrix $Q = (q_{\alpha\beta})$ is given as follows: Case (i):

$$q_{\alpha\beta} = 2 \operatorname{tr}(A_{\alpha}^{\Lambda}A_{\beta}^{\Lambda}),$$

where A_{α} is defined from A in (3.10) by substituting c into c_{α} . Case (ii)

$$q_{\alpha\beta} = 2 \operatorname{trA}_{\alpha}^{(1)} A_{\beta}^{(1)} + 4 \operatorname{trA}_{\alpha}^{(2)} A_{\beta}^{(2)} \Theta,$$

where $A_{\alpha}^{(1)}$ and $A_{\alpha}^{(2)}$ are defined from the $A^{(1)}$ and $A^{(2)}$ in (4.9) by substituting c into c_{α} .

$$q_{\alpha\beta} = \frac{1}{4} \operatorname{trA}_{\alpha} \begin{bmatrix} I_{p} & P \\ \\ P' & I_{q} \end{bmatrix} A_{\beta} \begin{bmatrix} I_{p} & P \\ \\ P' & I_{q} \end{bmatrix},$$

where A_{α} is defined from A in (5.14) by substituting c into c_{α} . Higher order terms of the joint distribution of J_1, \ldots, J_k can be obtained by using perturbation technique.

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Y.Fujikoshi, P.R.Krishnaiah, J.Schmidhammer	F49620-85-C-0008
PERFORMING ORGANIZATION NAME AND ADDRESS Center for Multivariate Analysis	10. PROGRAM ELEMENT, PROJECT, TASK AREA & WORK UNIT NUMBERS
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Air Force Office of Scientific Research Department of the Air Force	August 1985
Bolling Air Force Base, DC 20332	13. NUMBER OF PAGES
MONITORING ACENCY NAME & ADDRESS(II different from Controlling Office	e) 15. SECURITY CLASS. (of this report)
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